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Discrepancy between simulated and observed ethane and propane levels explained by underestimated fossil emissions

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Ethane and propane are the most abundant non-methane hydrocarbons in the atmosphere. However, their emissions, distribution in the atmosphere, and trends in their atmospheric concentrations are insufficiently understood. Atmospheric model simulations using standard community emission inventories do not reproduce available measurements in the Northern Hemisphere. Here, we show that observations of preindustrial and present-day ethane and propane can be reproduced in simulations with a detailed atmospheric-chemistry transport model, provided natural geologic emissions are taken into account and anthropogenic fossil fuel emissions are assumed to be two to three times higher than indicated in current inventories. Accounting for these enhanced ethane and propane emissions results in simulated surface ozone concentrations that are 5 to 13 % higher than previously assumed in some polluted regions in Asia. The improved correspondence with observed ethane and propane in model simulations with greater emissions suggests that the level of fossil (geologic + fossil fuel) methane emissions in current inventories may need re-evaluation.

Direct emissions at the surface are the only sources of ethane and propane to the atmosphere^{1,2}, and several studies suggest that they are underestimated in global inventories^{1,3-7}. A major source of uncertainty is that these inventories first calculate total non-methane hydrocarbon (NMHC) emissions and then disaggregate them into individual species (ethane, propane, etc.) based on limited amounts of data⁸⁻¹¹. Over the past decade the inventories do not fully account for an abrupt increase in the exploitation of unconventional natural gas in the United States^{3,4} and therefore likely underestimate present-day emissions^{3,4,12-23}. Recent atmospheric model simulations applying current global emission inventories tend to underestimate observed ethane and propane concentrations in wintertime in the Northern Hemisphere^{3-5,7,12,24-29}.

Anthropogenic fossil fuel (conventional and unconventional) emissions are presently the largest emission source of ethane and propane in most global inventories. Fugitive emission is the main fossil fuel NMHC source and includes venting and flaring, evaporative losses, and equipment leaks but not fuel combustion. Changes in these particular emissions are regarded as the main cause of observed ethane trends^{3,4,26,30-34}. Recent studies^{9,35} calculated fugitive fossil fuel emissions from oil, natural gas, and coal systems for ethane based on a joint inventory and

atmospheric box-model approach. Their emission dataset⁹ was recently updated with new data³⁶. Another recent study¹⁰ used a detailed inventory approach to identify cold venting of associated petroleum gas containing also methane, propane, and butane as significant emission sources potentially underestimated in existing emission inventories. These new studies combine field measurements and country-specific information from published sources along with observed flaring of associated gas from satellite images. In combination these take into account that the emission factors from venting and flaring of associated gas released during extraction vary considerably across different oil, coal, and gas fields around the world. Such considerations have not been made in most community emission datasets, which apply emission factors reported by countries to the United Nations Framework Convention on Climate Change, or from measurements representative for North America for those countries that are not reporting. Deficiencies in fugitive fossil fuel emission estimates in community emission datasets were also recently found for black carbon at high latitudes³⁷ and SO₂ at low latitudes³⁸ (the Middle East). Natural geologic emissions are another suggested fossil hydrocarbon source missing in inventories^{39,40}, receiving little attention in previous model studies. Major geologic sources include seepage from onshore and submarine petroleum basins, volcanoes, and degassing from geothermal manifestations (see Methods). Based on the few available estimates^{39,40}, geologic emissions may have been the largest preindustrial source of ethane to the atmosphere (Figure 1a).

Another cause of poor model performance could be inaccurate representation of atmospheric sinks. Oxidation by hydroxyl in the troposphere is the main sink for ethane and propane^{1,2,41}. Estimates of global mean hydroxyl levels and hemispheric ratios differ substantially between observation-based estimates and results from model ensembles⁴²⁻⁴⁴. Studies also find large inter-model differences in the atmospheric distributions of oxidants⁴⁵.

In this study we first apply the OsloCTM3 model⁴⁶ to investigate the preindustrial atmospheric ethane budget. We compare model results from simulations with and without geological emissions to ice-core measurements. Thereafter, we model current conditions represented by the year 2011, which is the last year available in all applied fugitive fossil fuel emission datasets. In the year 2011 baseline simulation we use the state of the art global anthropogenic emission inventory CEDS CMIP6⁴⁷. We also include natural emissions (treated as negligible in many

model studies) from oceanic, biogenic (vegetation) and geologic sources (see Methods). We compare with surface ethane and propane measurements from global and regional surface networks (see Methods) with a focus on observations in the mid- to high-latitude Northern Hemisphere where previous model studies underestimated observations. We then generate alternative gridded emissions by replacing the fugitive fossil fuel emissions in the CEDS CMIP6 inventory with new datasets that better represent fossil fuel activity and emissions^{9,10,35,36} with corrections to avoid double counting from potential overlap with natural geologic emissions (see Methods). In the resulting alternative simulations (ALT1 and ALT2, Tables 1-2) the fossil fuel emissions are factors of about two (ethane, Figure 1b) and three (propane, Figure S1) higher than in the baseline simulation with very different geographical distributions (Figure S2). We also suggest further modifications of the ALT1 and ALT2 emissions based on studies with the Flexpart model⁴⁸. Back-trajectories from Flexpart are used to identify source regions related to OsloCTM3 under- and overestimation of observed ethane concentrations (see Methods). Finally, we explore the uncertainty of the atmospheric sinks in a sensitivity study where we perturb the OsloCTM3 hydroxyl level within its uncertainty range (see Methods).

Preindustrial ethane budget

Ice core measurements³⁹ reveal a large, factor of about 4, north/south inter-polar ratio for ethane. Figure 2a shows that this can be reproduced by the model with a geologic source of 3 Tg/yr, constituting about 40% of the total preindustrial ethane emissions (Figure 1a). With geologic emissions included in the simulations the modeled abundance at Summit in Greenland agrees with observations, and the simulated inter-polar ratio improves substantially relative to observations. Without geologic sources the simulated abundance at Summit is 50 % too low. This is in agreement with the findings of ref. 39, where a simpler model without interactive oxidation chemistry was used.

Our baseline simulation slightly overestimates the observed Antarctic ethane concentration. Transport to high southern latitudes and biomass burning emissions have high inter-annual variability. However, sensitivity simulations with meteorological input data for a different year and an alternative inventory with different geographical distribution and emission totals for biomass burning emissions (see Figure 1a and Methods) resulted in minor changes (see

Supplementary). We therefore suggest that a small Antarctic overestimation relates to uncertainties in the magnitude and distribution of geologic and oceanic emissions in the Southern Hemisphere.

Current ethane and propane budgets

The baseline simulation for 2011 does not reproduce the observed inter-polar ethane ratio well, even if geologic emissions are included (Figure 2b). The modeled concentration at Summit in Greenland is only about 50 % of measured values (Figure 2b). Underestimations of ethane and propane concentrations at high northern latitudes, particularly during wintertime (Figure 3, Figures S3-S7), are similar to most other model studies^{3-5,7,12,24-29} using standard emission inventories. As shown below, the likely cause is underestimated fossil fuel emissions in the standard community emission data set CEDS CMIP6, used in the baseline simulation. The CEDS CMIP6 emission data⁴⁷ agree with previous emission data for ethane and propane used in atmospheric climate and air pollution studies (Figure 1b and Figure S1).

The ALT1 and ALT2 simulations, where the CEDS CMIP6 fugitive fossil fuel emissions are replaced with the new datasets^{9,10,35,36}, reproduce the inter-polar ethane ratio and the observed levels in Greenland (Figure 2b), Zeppelin Observatory at Svalbard (Figure 3a), and most other stations (Figure 3b, Figures S3-S9) much more closely. This is also the case for propane (Figure 3a, c, Figures S3-S9), for which fossil fuel emissions play an even larger role (Figure S1). A substantial improvement is found throughout the Arctic (Figure S5, Tables S2-S3). ALT1 performs better than ALT2. Both have positive mean biases, tending to overestimate episodes with high concentrations. We explore these and other episodes at Zeppelin (Arctic station with frequent sampling) in a systematic way (see Methods). Figure 4 shows that the episodes with the largest underestimation of ethane at Zeppelin in the baseline simulation occur for air masses originating from Eurasia. Fossil fuels are the dominant emission source in this region for most of the year, strongly suggesting that these are underestimated in the CMIP6 inventory. From Figure 4, it is also evident that ALT1 underestimates the fossil fuel emissions in northwestern Europe (i.e. mainly emissions from the North Sea) while it overestimates emissions from Russia. The ALT2 simulation mainly overestimates observed ethane levels at Zeppelin (Figure 3a). For this inventory, the fossil fuel emissions are likely overestimated both in the North Sea and Russia

(Figure 4). Overestimated emissions in ALT1 and ALT2 over Russia also seem likely from the comparison with Tiksi station data (Figure S5, the only available station in Russia).

At mid-latitude stations in the U.S. and Canada, both ALT1 and ALT2 show good agreement with measurements (Figure S6, Tables S2-S3). An exception is the highly oil and natural gas influenced Southern Great Plains station (large underestimation) (Figure S6). Large emissions from several nearby oil wells might not be fully resolved in the model averaging emissions over the model grid scale. It is also a possibility that the ALT1 and ALT2 inventories underestimate the emissions from nearby unconventional gas fields (e.g. Woodford, Barnett⁴⁹) and oil wells. The ALT1 and ALT2 simulations also improve agreement with measurements at non-Arctic European stations compared to the baseline (Figure S7, Tables S2-S3), especially for ethane. At most stations the ALT1 simulation is biased slightly low compared to the observations. ALT2 also performs better than the baseline, but overestimates, to varying degree, the measurements at most European stations.

The lower latitude Cape Verde site also shows large improvements (ALT simulations vs. baseline) during wintertime (Figure S8) when it is influenced by air passing over the Sahara⁵⁰ from oil and gas fields in northern Africa and the Middle East. In the Southern Hemisphere, the baseline simulation reproduces observed levels and seasonal patterns well and the alternative simulations only result in minor differences (Figures 2-3 and Figure S9).

The ALT1 and ALT2 anthropogenic ethane emissions (excluding biomass burning) are slightly smaller than the optimized anthropogenic emissions in other recent model studies^{3,4} (Figure 1b). The optimized emissions in other studies are based on sensitivity simulations^{3,4} finding that an approximate doubling of anthropogenic emissions is needed to reproduce measurements at Jungfrauoch⁴ and a few other FTIR stations in the Northern Hemisphere³. In our study we include natural geologic emissions and apply new detailed emission datasets for fugitive fossil fuel emissions instead of performing an up-scaling of all anthropogenic emissions. Our model is also run at higher spatial and temporal resolution and compared to a larger number of measurement sites.

Overestimated atmospheric loss (i.e. too high hydroxyl levels) might lead to an underestimation of observed ethane and propane levels. However, this cannot be a major cause for the discrepancies in our baseline simulation. Scaling down tropospheric hydroxyl levels to the lower

range of model and observational based estimates (see Methods) improves the agreement slightly (Figure S10) but much of the underestimation in the mid-high Northern Hemisphere during wintertime persists. This seasonal pattern is expected as chemical loss at high northern latitudes is inefficient during wintertime when little sunlight and low water vapor concentrations result in low hydroxyl concentrations.

We have not included atmospheric oxidation of ethane and propane by halogens in the model simulations. The reasons and implications on uncertainty are discussed in the Supplementary. Inclusion of halogen chemistry would likely lead to slightly larger underestimation of ethane and propane in the baseline simulation, thereby supporting that emissions in standard community datasets are too low.

Our global total natural geologic emissions correspond with the best estimates from the study by ref. 40. Reported uncertainty ranges^{39,40} are shown in Figure 1 and Figure S1. A recent study⁵¹ suggests geologic methane emissions about 1/3 of that estimated by ref. 40. NMHC to methane emission ratios from geologic sources are uncertain and likely highly spatially and temporally variable. Therefore, findings suggesting lower methane flux in the far past do not necessarily imply lower preindustrial and present-day NMHC emissions. However, if the finding⁵¹ was applied via downscaling to estimate ethane emissions, it then fits poorly with the findings of our preindustrial simulations (Figure 2a), unless we change the geographical distribution towards a larger fraction of emissions in the Northern Hemisphere. As noted (see Methods) the uncertainty in geographical distribution is large.

The relative uncertainties in global total emissions in the new fugitive fossil fuel datasets (applied in the ALT1, ALT2 inventories) are about half of those used in CEDS CMIP6 (baseline inventory) (see Supplementary). Another improvement is substantially reduced uncertainties in geographical emission distributions (see Methods). Based on the comparisons in the previous paragraphs the correct global total anthropogenic ethane emissions seem to be close to the levels in the ALT1 and ALT2 simulations (rectangle, triangle Figure S11). These levels are greater than the upper cap of the baseline uncertainty bar in Figure S11, i.e. the baseline inventory likely underestimates emissions. The total emissions in the baseline inventory are close to the mean and median of those in eight other standard community emissions datasets suggesting that applying

these will also result in too low modeled ethane concentrations. For the ALT1 and ALT2 inventories, only total emissions near the lower end of the uncertainty range (Figure S11) reproduce the observed levels. Various model uncertainties widen the possible emission range but major ones associated to the OH sink have relatively small impacts on modeled ethane concentrations. Model uncertainties do therefore not change our conclusions regarding under- and overestimation in the different emission inventories. Using the alternative emission datasets (ALT1 or ALT2) instead of the standard community emission inventories greatly improves the comparison with observations. Due to sparse observation coverage in some world regions, and the uncertainty ranges of the emissions and atmospheric chemical loss of ethane and propane, we do not provide an overall performance ranking between the emission datasets ALT1 and ALT2.

Impacts on other atmospheric constituents

The higher ethane and propane in the ALT1 and ALT2 simulations compared to the baseline simulation impact the greenhouse gas methane and major surface pollutants. The impacts on tropospheric methane are moderate, leading to 0.5 % (ALT1) and 0.7 % (ALT2) higher methane lifetimes due to lower tropospheric hydroxyl. Modeled baseline ozone mixing ratios are compared to surface measurements in Figure S12 for the period June-August when ozone photochemistry is most active in the Northern Hemisphere. The model reproduces the gradients between regions with high photochemical production and cleaner background areas. At many stations the model is at, or within, a few ppb (or percent) of the measurements. In regions with high levels of surface ozone in the baseline, particularly the Middle East and eastern Asia, ozone is 5-13 % (3-11 ppb) higher in spring/summer (Figure S13) in ALT1 and ALT2. In these regions, the ozone production is more sensitive to the amount of NMHCs since high NO_x is present. Surface ozone differences in other regions are generally small (0-5 % or 0-3 ppb). If fossil emissions of other related NMHCs are underestimated as well (e.g. butane, pentane etc.), the impacts on ozone and other air pollutants will be larger. Impacts on the air pollutants NO₂, PAN and CO are discussed in the Supplementary.

Methane constitutes the largest share of hydrocarbons emitted from fossil sources, and a recent study⁵² suggests underestimation of fossil (geologic + fossil fuel) methane emissions in previous estimates. Compared to previous inventories the much higher fossil fuel ethane and propane emissions in the new datasets^{9,10,35,36} used in ALT1 and ALT2 in this study are mainly due to higher NMHC to methane emission ratios. The improved agreement with ice-core ethane measurements for the simulation with geologic emissions supports the idea that there is a considerable geologic methane emission source^{52,53}. As for ethane and propane, geologic emissions of methane have been neglected in many model studies. In accordance with ref. 52 we suggest a need for more studies evaluating the reported level of fossil (natural geologic + anthropogenic fossil fuel) methane emissions in current emission inventories. Understanding the contribution from different natural and anthropogenic emission sources is a critical precursor to design efficient measures to reverse ongoing atmospheric ethane, propane, and methane increases^{4,54,55}.

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Author contributions

S.B.D., G.M., and Ø.H. designed the study with input from A.S., C.L.M., and I.P. S.B.D. performed the simulations with the OsloCTM3 model, analyzed the model results, and performed the comparisons with measurement data. Ø.H. and G.M. provided assistance with the analysis and comparison studies. I.P. performed the simulations with the Flexpart model and I.P. and A.S. analyzed the output. S.SC. and L.H.I. provided the new emission datasets for fugitive fossil fuel emissions. S.B.D. developed gridded inventories for geologic emissions. C.L.M, D.H., S.R., S.SA., N.S., K.A.R., L.J.C., A.C.L., S.P., and M.W. provided the observational data for ethane and propane. S.B.D. led the writing of the manuscript in close collaboration with G.M. and Ø.H. All authors contributed to the writing and review of the manuscript.

Competing financial interests

The authors declare no competing financial interests.

Figure captions

Figure 1: Global total sectoral ethane emissions in this study and other studies. a) Global total ethane emissions in preindustrial simulations in this study (black symbols) compared to emissions in other studies (blue and yellow symbols and bars). *See Methods section for details. ^δBox model study³⁹. [†]Bottom up inventory⁴⁰. **b)** Global total ethane emissions in the year 2011 baseline and alternative (ALT1, ALT2) simulations in this study (black symbols) compared to emissions in other studies. The inventories and estimates cover the year 2000 and onwards. The closest year to our simulation year 2011 is chosen for inventories not covering that year. Bottom up inventories (yellow symbols and bars)[#]: Fossil fuel, biofuel, agriculture, waste: CEDS CMIP6⁴⁷ (used in baseline in this study), HTAPv2, Edgar 3.2 FT, RETRO, POET, CMIP5 (average of MACCITY, ACCMIP, RCP2.6, RCP4.5, RCP6, RCP8.5) (as reported and referenced in ECCAD database: <http://eccad.aeris-data.fr/>), ARCTAS (as reported by ref. 7), EDGAR4.3.2 (as reported by ref. 11), and new inventory in ref. 56. Biomass burning: GFEDv4 (as used in baseline in this study), GICC, ACCMIP, POET, GFASv1.2, MACCITY, RETRO, RCP2.6, RCP4.5, RCP8.5 (as reported and referenced in ECCAD database: <http://eccad.aeris-data.fr/>), and FINN (as reported by ref. 3,4). Oceans: RETRO (used in baseline in this study). Vegetation: MEGAN-MACC (used in baseline this study), and MEGANv2 (as reported and referenced in ECCAD database: <http://eccad.aeris-data.fr/>). Natural geologic: As reported by ref. 40 (median estimate used in baseline in this study). Top-down estimates from box models (dotted blue bars)^Φ: Fossil fuel, biofuel, agriculture, waste: ref. 31,33,39. Biomass burning: ref. 39. Geologic: ref. 39. Optimized emissions in 3D model studies (brown symbols)*: Ref. 3,4.

Figure 2: Observed and modeled annual mean ethane mixing ratios and inter-polar ratios. a) Observed³⁹ and modeled (this study) preindustrial inter-polar ratio and mixing ratios at Summit, Greenland and at the West Antarctic Ice Sheet (WAIS) site. Observation error bars are the reported ± 2 standard errors in ref. 39. **b)** Observed and modeled inter-polar ratio and mixing ratios in 2011. For the Antarctic, the closest station (South Pole) with data to the WAIS measurement site (no data for 2011) was used. See Table 1 for more information about the simulations.

Figure 3: Comparison of year 2011 modeled and observed ethane and propane at surface sites. **a)** Comparison of modeled and observed year 2011 ethane (upper row) and propane (lower row) at the Zeppelin station. A selection of comparisons for other sites is shown in the Supplementary. **b)** Comparison of modeled (background colors) and observed surface (color-filled circles) of ethane (ppt) for the year 2011. Model data for the lowest model layer were used. Stations with less than 6 samples within the 3 months averaging period were excluded from the comparison. Mountain stations typically sampling free tropospheric air and situated in areas where the model resolution does not resolve the terrain were also excluded. Details on the applied observation datasets are provided in the Methods section. Maps for the ALT2 simulation are shown in the Supplementary. **c)** Same for propane.

Figure 4: Footprints at Zeppelin. Yearly mean Flexpart footprints (see Methods for details on approach) for ethane at Zeppelin. Upper left: Baseline simulation. Episodes with Largest Underestimation (LU). Upper right panel: ALT1 simulation. Episodes with LU. Lower left: ALT1 simulation. Episodes with Largest Overestimation (LO). Lower right: ALT2 simulation. Episodes with LO. The unit ns/kg is proportional to the residence time in a given volume of air.

Methods

Models:

OsloCTM3:

We use the OsloCTM3 model⁴⁶ to simulate the preindustrial (year 1750) and current (year 2011) distributions of atmospheric ethane and propane. The model is run with approximately $1.1^{\circ} \times 1.1^{\circ}$ (T159) horizontal resolution. To spin up the model 15 months simulations were made with coarse resolution ($2.2^{\circ} \times 2.2^{\circ}$) followed by 4 months with $1.1^{\circ} \times 1.1^{\circ}$ (T159) resolution. After the spin up a set of simulations (Table 1) were made. A coupled tropospheric and stratospheric 60 layer (surface-0.1 hPa) version is used with 100 chemical active species affecting atmospheric oxidation capacity. OsloCTM3 was described and evaluated by ref. 46 and was used in several studies related to atmospheric oxidation capacity⁵⁷.

The OsloCTM3 simulations are driven with 3-hourly year 2011 meteorological forecast data from the European Centre for Medium-Range Weather Forecasts (ECMWF) Integrated Forecast System (IFS) model (see ref. 46 for details). These data are 36-hours forecasts produced with 12 hours of spin-up starting from an ERA-Interim analysis at noon on the previous day.

Flexpart:

To investigate the origin of air masses observed at the Zeppelin station, we use version 9.2 of the LPDM FLEXPART (FLEXible PARTicle dispersion model)⁴⁸. The model is driven with 3-hourly operational meteorological analyses from the European Centre for Medium-Range Weather Forecasts (ECMWF) with 91 vertical levels and a horizontal resolution of $1^{\circ} \times 1^{\circ}$. Computational particles released from the location of the Zeppelin station are tracked 20 days back in time in FLEXPART's "retroplume" mode. The model output consists of an emission sensitivity, the surface footprint of which is used here to identify source regions related to OsloCTM3 under- and overestimation of observed ethane concentrations.

Emissions and model simulations

Baseline emission inventories all constituents

For anthropogenic SO_x, NH₃, CO, NO_x, and NMHCs emissions, we use the Community Emissions Data System (CEDS) Project emission inventory⁴⁷ for the years 1750 and 2011. CEDS is the state of the art dataset currently used in the Coupled Model Inter-comparison Project Phase 6 (CMIP6). For biomass burning, we use GFEDv4⁵⁸ year 2011 emissions and the historical biomass burning dataset for CMIP6 for 1750⁵⁹. Sulfur emissions from other sources are described in ref. 60. Non-methane volatile organic compounds (NMVOCs) emissions from vegetation and oceans are neglected (set to zero) in some studies but not in this study. Biogenic emissions of CO and NMVOCs from vegetation are set to year 2010 (last year covered by dataset) emissions from MEGAN-MACC⁶¹ both for the preindustrial and year 2011 simulations. For NO_x from soil and CO and NMHCs from the oceans we use the year 2000 emissions in the RETRO inventory⁶². Lightning NO_x emissions are described in ref. 46. For natural NH₃ sources we use emissions from GEIA⁶³ for 1990. Methane emissions are described in ref. 57.

Geologic emissions of ethane and propane

For ethane and propane we include geologic emissions suggested by ref. 40 in the baseline emission inventories. For both 1750 and 2011 the global total emissions are set to the medians (3 Tg/yr for ethane, 1.7 Tg/yr for propane) of the ranges (2-4 Tg/yr ethane, 1-2.4 Tg/yr propane) estimated by ref. 40. Their study splits the emissions into six main geologic sources: Mud volcanoes, gas seeps, microseepage, submarine seeps, geothermal manifestations and volcanoes. The geographical distribution of geologic emissions has not been gridded to files suitable for atmospheric modeling studies. Here we use several datasets to develop gridded inventories. The emissions from gas seeps (macroseepage) and mud volcanoes are distributed in accordance with the GLOGOS dataset which lists more than 2000 terrestrial (onshore) seeps from 87 countries. In addition to site locations, GLOGOS provides measured or estimated (visually) fluxes of methane for a few sites and methane, ethane and propane concentrations in the gas for some more sites. However, the majority of sites lack such information and we therefore distribute the emissions evenly on the sites to obtain the global total macroseepage emissions estimates from ref. 40. For submarine seeps we used a derived product (see acknowledgement) from the Global Offshore Seepage Database (GOSD) indicating where offshore seepage occurs. We scale the density map from this dataset to obtain the global total emissions from marine seepage in ref. 40. We assume zero emissions from marine seepage in grid-boxes fully covered by sea-ice. For

emissions from microseepage we use the CGG Robertson Tellus Sedimentary Basins of the World Map to distribute the emissions from ref. 40. Microseepage, which is diffuse exhalation from soil in petroleum basins, is the largest geologic source but also the most uncertain one^{40,64} in terms of magnitude and geographical distribution. We spread the emissions evenly over the world's petroleum basins, which represents the potential area for such diffusion and we likely to some degree overestimate the geographical extent. We assume that permafrost or thick ice- and snow-layers hinder diffusion. The northward and southward extent of emissions is therefore limited to 66° N and 60° S to account for this in a simplified way. The emissions from volcanoes and geothermal sources are gridded using the geographical distribution for SO₂ emissions⁶⁰ for such activity. Geologic seepage (macroseepage, some microseepage, and marine seepage) occur at many of the places where current oil, gas, and coal extraction take place.

Baseline and alternative year 1750 preindustrial ethane emissions and simulations

The preindustrial baseline simulation (Table 1) includes geologic emissions of the magnitude suggested by Etiope et al.⁴⁰ and within the range found by Nicewonger et al.³⁹ (Figure 1a). In Nicewonger et al. an observed inter-polar ethane asymmetry requires a certain combination of emissions from biomass burning and natural geologic sources (Figure 1a). Preindustrial biomass burning emissions are particularly uncertain, and the magnitude in the inventory (CEDS CMIP6 year 1750) in our baseline simulation is high compared to the range obtained by ref. 39. We therefore perform an additional simulation using a biomass burning inventory (CMIP5 1850⁶⁵) with lower emissions (Table 1, Figure 1a) and different geographical distribution. Transport to high southern latitudes has inter-annual variability. To check the sensitivity on our results we do a simulation with meteorological input data for a different year (Table 1).

Baseline year 2011 ethane and propane emissions and simulation

Figure 1b shows global total sectoral ethane emissions used in the year 2011 baseline simulation in this study compared to emissions used in other inventories. Based on the preindustrial simulations we include geologic emissions in our baseline simulation for current conditions (year 2011). Except for the box model optimized inventory for 2000-2010 in Nicewonger et al.,

geologic emissions were not included in any other model studies. Our applied emissions for biomass burning (GFEDv4 year 2011) are in the mid-range compared to other inventories. The other anthropogenic emissions (sum of fossil fuel, biofuel, agriculture, and waste in Figure 1b) in the CEDS CMIP6 data⁴⁷ applied in this study are quite close to the median of the estimates in other inventories. Due to likely trends in anthropogenic emissions over the last decades^{3,4,31,33}, different basis years partly explain the large range of the emission estimates in various inventories presented in Figure 1b. The other reason is the large uncertainties in existing inventories due to incomplete approaches and datasets (discussed in main text and section below). Summing up all sectors in Figure 1b our total emissions are in the upper range of other studies. This is mainly due to the inclusion of geologic emissions.

Alternative year 2011 ethane and propane emissions and simulations

We also perform simulations with alternative ethane and propane emissions from the energy sector (Table 1). In these we replace emissions from the energy sector in the CEDS CMIP6 inventory with two new datasets^{9,10,35,36} for upstream (fuel production, gathering, and processing) and downstream (transmission and distribution) emissions from oil, gas and coal systems. These studies take much better into account that the emission factors from venting and flaring of associated gas released during extraction vary considerably across different oil, coal and gas fields in the world. The studies used novel approaches to quantify and attribute methane and NMHC emissions, combining field measurements and country specific information from published sources with observed flaring of associated gas from satellite images, to arrive at country-specific emissions from flows of associated gas. Two simulations are performed with these datasets. In what we refer to as the ALT1 simulation emissions from oil and gas from ref. 10 are combined with coal emissions from ref. 9 (updated with data from ref. 36) to obtain a complete inventory for the energy sector. In the simulation entitled ALT2 we use oil, gas, and coal emissions from ref. 9,35 updated with data from ref. 36. The dataset does not include propane emissions and we use the global mean propane to ethane emission ratio from the ref. 10 datasets to obtain propane emissions for the ALT2 simulation. In ALT2 we use the low estimate for natural gas from ref. 9 since it was shown to be the most likely³⁵. Table 2 provides an overview of the fugitive fossil fuel emissions in the ALT1 and ALT2 inventories. Due to

substantial geographical overlap it is likely that some emissions from geologic seepage are included³⁵ in fugitive fossil fuel inventories. To avoid double-counting, we subtract ethane and propane emissions from the oil, gas, and coal grids. In the absence of a well-established gridded emission inventory of geologic seeps, we subtract geologic seepage ethane and propane emissions from oil, gas, and coal grids in equal parts, i.e., one third each. By reducing with amounts corresponding to all emissions from geologic seepage, the resulting inventories (ALT1 and ALT2) could be regarded as lower estimates of emissions from current oil, gas, and coal activity relative to ref. 10 and ref. 9.

Uncertainties in baseline and alternative anthropogenic emission inventories

The uncertainty calculations for the baseline inventory and alternative (ALT1, ALT2) inventories are discussed in the Supplementary in relation to Figure S11.

Sensitivity simulation on atmospheric sink

Oxidation by hydroxyl (OH) in the troposphere is the main sink for ethane and propane^{1,2,41}. The uncertainty for the reaction rates are rather small, 15-20 % at 298 K^{2,66}. For hydroxyl concentrations, the uncertainty is larger. Our global averaged tropospheric OH (1.35×10^6 molecules/cm³) in our 2011 baseline simulation is on the high side compared to other model studies⁴⁵. The same holds for the hemispheric OH ratio (1.55) compared to model and observational based estimates^{42,44}. The modeled global average methane lifetime, which is highly dependent on the modeled OH concentration, is also low compared to observational based estimates^{67,68}. We therefore did a sensitivity study scaling down the global mean OH concentration to 1×10^6 molecules/cm³. The scaling was done separately for the hemispheres to also get a hemispheric ratio of 1. This is in the lower range of model and observation based estimates both for the OH concentration and hemispheric ratio and was done to see to what degree lower OH concentrations in the Northern Hemisphere would improve the comparison with observations. The OH concentration was scaled down only in the chemical reactions with ethane and propane and not in reactions with other atmospheric constituents.

Measurement data

More and more high quality measurements of NMHC have become available through coordination by the World Meteorological Organization (WMO) Global Atmospheric Watch (GAW) program. Participating networks include ACTRIS (Aerosol, Clouds, and Trace gases Research Infrastructure, the European Research Infrastructure for the observation of Aerosol, Clouds, and Trace gases (<http://www.actris.eu>)), EMEP (The European Monitoring and Evaluation Program) and NOAA ESRL/INSTAAR (National Oceanic & Atmospheric Administration - Earth System Research Laboratory/Institute for Arctic and Alpine Research). To reveal strengths and discrepancies in model performance and evaluate emission inventories we compare the model results to surface ethane and propane observations for the year 2011. We use data from surface sites reported to the World Data Center for Greenhouse Gases (WDCGG) (<http://ds.data.jma.go.jp/gmd/wdcgg/>) following the NOAA/INSTAAR and WMO-GAW measurement protocols, and from EMEP complying with ACTRIS recommendations. EMEP and ACTRIS data were downloaded from EBAS (<http://ebas.nilu.no>), and are also accessible through the ACTRIS data portal (<http://actris.nilu.no>). These data follow the ACTRIS quality assurance procedures and protocols. Generally, the ACTRIS data have higher time resolution (up to 2 hour). Intercomparison exercises⁶⁹⁻⁷¹ have shown consistency of NOAA/INSTAAR and EMEP/ACTRIS measurement sites within the data quality objective of $\pm 10\%$ of the WMO Global Atmospheric Watch program^{72,73}.

A subset of data from 96 sites from a total of 132 sites (66 for both ethane and propane) are shown in the comparisons for 2011. The observations at the given location, altitude and time were compared to output for the closest model grid box, level, and time. Figure 3 shows all sites and Figure S4 shows the locations of the subset of sites selected for detailed comparison to model results. Criteria for selection were data quality assurance, access to continuous time series with few gaps, coverage of different regions, and site characteristics (e.g. elevation, topography, and influence of pollution episodes) likely captured by the resolution of a global model.

A comparison between modelled and observed ozone is performed in the Supplementary and the applied ozone measurement data are presented there.

Table 1: Overview of the simulations performed with OsloCTM3.

Year	Simulation	Simulation name	Inventories anthropogenic and natural emissions*	Geologic emissions	Oxidation chemistry	Meteorological year
1750	Preindustrial	1750 Baseline	Baseline	Yes	Interactive	2011
	Preindustrial sensitivity	1750 NOGEO	Baseline	No	Interactive	2011
		1750 MET	Baseline	Yes	Interactive	2014
		1750 BBURN	Baseline, but CMIP5 biomass burning	Yes	Interactive	2011
2011	Baseline	2011 Baseline	Baseline	Yes	Interactive	2011
	Baseline sensitivity	OH	Baseline	Yes	Interactive, OH field scaled down in reactions with ethane and propane**	2011
	Alternative	ALT1	ALT1	Yes	Interactive	2011
		ALT2	ALT2	Yes	Interactive	2011

*The baseline, ALT1 and ALT2 emission inventories are described in the above section.

** The scaling of OH is described in the above section.

Table 2: Overview of the new fugitive fossil fuel emission inventories used in the alternative year 2011 simulations.

Inventory	Natural gas	Oil	Coal
ALT1 ethane	ref. 10	ref. 10	ref. 9 updated with data from ref. 36
ALT1 propane	Same as above	Same as above	Same as above
ALT2 ethane	ref. 9,35 updated with data from ref. 36	ref. 9 updated with data from ref. 36	ref. 9 updated with data from ref. 36
ALT2 propane	Same as above using propane to ethane ratio from ref. 10	Same as above using propane to ethane ratio from ref. 10	Same as above using propane to ethane ratio from ref. 10

Code availability. We have opted not to make the computer codes associated with this paper available, because replication of our results does not require access to the computer codes.

Data availability. The ethane and propane surface measurement data used in this study are freely available. We use data reported to the World Data Center for Greenhouse Gases (WDCGG) (<http://ds.data.jma.go.jp/gmd/wdcgg/>). EMEP and ACTRIS data are available from EBAS (<http://ebas.nilu.no>), and are also accessible through the ACTRIS data portal (<http://actris.nilu.no>). The sites used for detailed comparison with model results are listed in Table S1. The sites can easily be found by name or map search in the databases. The new emission datasets for fugitive fossil fuel emissions are available upon request to Lena Höglund-Isaksson (hoglund@iiasa.ac.at) and Stefan Schwietzke (stefan.schwietzke@noaa.gov). The gridding used to develop geologic ethane and propane emissions suitable for atmospheric modeling studies was based on commercial datasets owned by CGG geoconsulting. CGG should be contacted and consulted for the task of gridding geologic emissions.

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